







PASSIVE AND ACTIVE CONTROL FOR REHABILITATION ROBOTS

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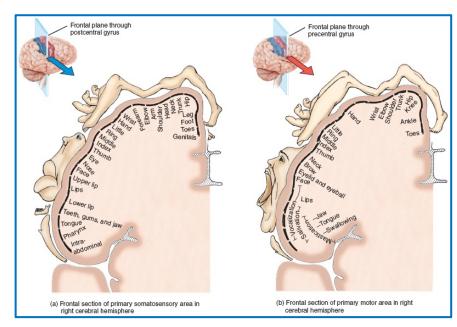
- Background
- Review of Rehab Robotics Research
- Research on Upper-limb Rehab Robot
- Research on Lower-limb Rehab Robot

- Background
 - Neurological Injuries: Stroke and SCI (Spinal Cord Injury)
 - Motor Plasticity and Rehabilitation
 - Manual Training VS Robot-aided Training
- Review of Rehab Robotics Research
- Research on Upper-limb Rehab Robot
- Research on Lower-limb Rehab Robot

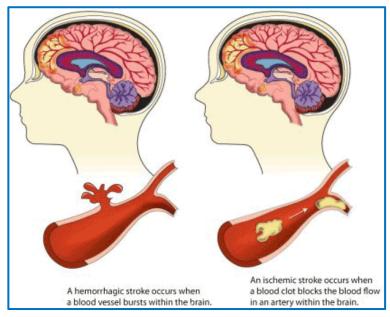
Neurological Injury

-Stroke

• The combination of sensory and motor deficits each person will experience after stroke vary greatly based on the location in the brain.



Sensory and Motor Cortex

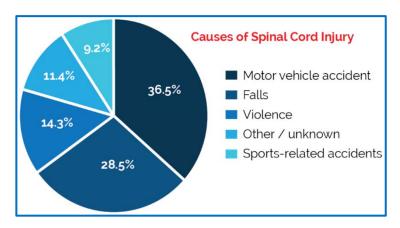


Hemorrhagic Stroke

Ischemic Stroke

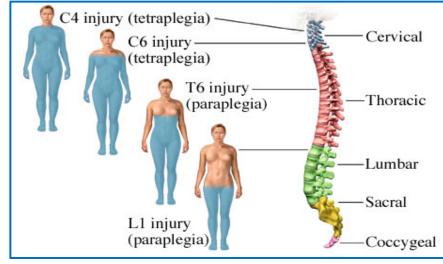
Neurological Injury

— Spinal Cord Injury (SCI)



- * Data from National SCI Statistical Center (NSCISC)
- Injuries can occur at any level of the spinal cord and can be classified as complete injury, a total loss of sensation and muscle function, or incomplete, meaning some nervous signals are able to travel past the injured area of the cord.

 SCI is a damage to spinal cord that causes loss of muscle function, sensation, or autonomic function in parts of the body served by the spinal cord below the level of the lesion.



Paralysis Based on the Location of Injury

- Background
- Review of Rehab Robotics Research
 - Related Research Areas
 - State-of-the-art Design
 - Training Mode
 - sEMG Processing and Modeling
- Research on Upper-limb Rehab Robot
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Areas Covered by Rehabilitation Robotics:

- Rehabilitation and assistive robotics
- Wearable assistive devices
- Behavioral neuroscience
- Biologically-inspired assistive systems
- Human-machine interaction
- Neuro-robotics
- Prosthetic devices
- Locomotion and manipulation assistance
- Technology assessment, ethical and social implications
- ...

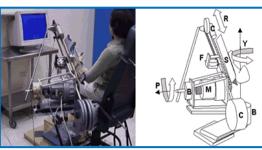
Upper Limb Rehabilitation Robot



MIT-MANUS, MIT, 1992



MIME, Stanford, 2000



ARM guide, U.C. Irvine, 2000



ARMin, ETH, 2005

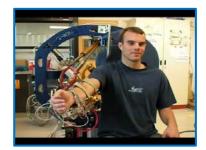
1992

2000

2005

2008

2007



BONES, U.C. Irvine, 2008



Gentle/S, EU, 2007



CADEN-7, UCLA, 2007



Pneumatic T-WREX, U.C. Irvine, 2007

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Lower Limb Rehabilitation Robot



Locomat, Swiss, 2004



ReoAmbulator, USA, 2004



MotionMaker, Swiss, 2004



LOPES, University of Twente, Netherland, 2007



HapticWalker, Germany, 2008

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Robot-aided Training Mode

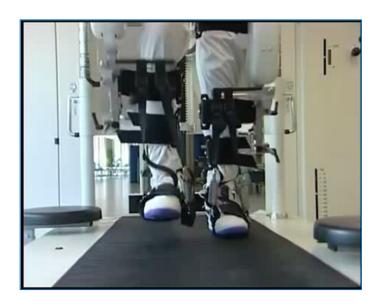
- —How the rehabilitation robot interacts with participants during training?
- Passive Training —robots provide assistance to move, and patients' active efforts are not considered explicitly in movement control
- Active Training —robots provide limited assistance or add resistance according to the participant's motion intention
- Assist-as-Needed Training —robots provide as much assistance/resistance as necessary adaptively based on the participant's performance

Mode 1: Passive Training

Positive effects:

- Helps move the affected limb in a manner that self-generated effort cannot achieve.
- Suitable for acute/sub-acute period poststroke and complete SCI patients who have very weak muscle force.
- Stretching helps prevent stiffing of soft tissue and reduce spasticity.
- Helps increase Range of Motion (ROM).





Example: Gait Training Using Locomat

On the other hand, passive training mode lacks human active engagement and enough sensorimotor stimulation, and therefore is less effective for motor recovery.

Mode 2: Active Training

Active Training

- To provide assistance/resistance according to the participant' s voluntary motion or motion intention.
- To promote motor relearning through the participant adjusting actively according to force, visual stimulations and feedbacks.
- There are still no general approaches for reliably extracting motion intention and providing assistance in a optimal manner.



Training using ARMin

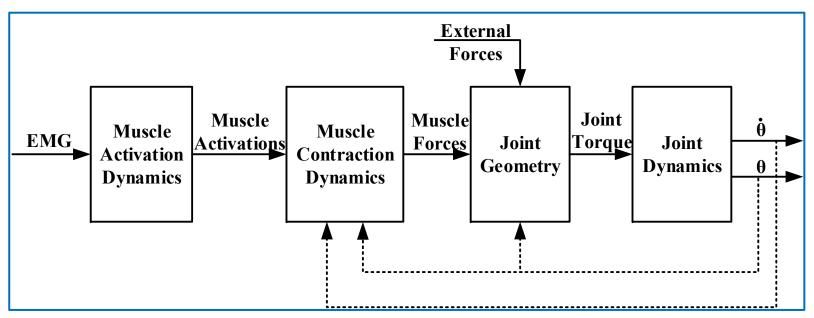
Too little assistance causes frustration and decreases motivation, whereas too much assistance decreases active output and encourages slacking.

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sEMG Processing and Modeling

-Neuromusculoskeletal Model Method

- To build a computational model between sEMG and muscle force/torque.
- All subsystem dynamics are subject-dependent, and muscle physiological parameters need to be identified.



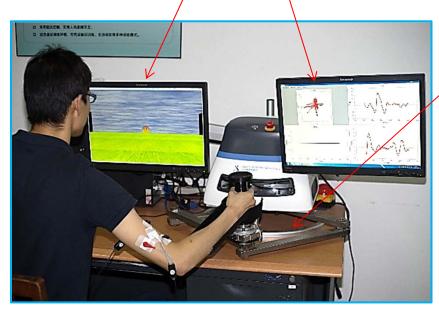
sEMG-driven Neuromusculoskeletal Model

- Background
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 - > Robot Design and Force Feedback
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Upper-limb Rehabilitation Robot

Virtual Training Environment Haptic Interface Force Feedback

Visual/Audio Feedback



Robot-aided Training Scenario

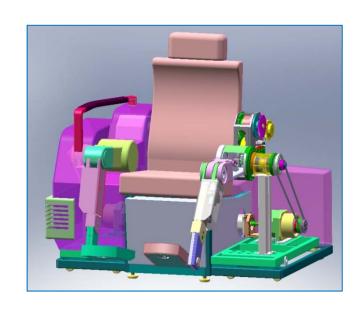


CASIA-ARM Rehab Robot

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 - Robot Design and Integrated Modules of sEMG and FES
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Lower-limb Rehabilitation Robot





Features:

- Three DOFs (thigh, knee, and ankle joints)
- Designed for patients with serious or minor spinal cord injury.
- For patients of height from 1.5m ~1.9m with length adjusting mechanism.

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Lower-limb Rehabilitation Robot

—Rehabilitation Training Mode

- By combing the rehabilitation robot and FES, sEMG devices, different training modes can be implemented:
- 1. Passive training: The patient's affected limb is moved by the robot to follow a predefined trajectory.
- FES-assisted training: During the passive training, major muscles are electrically stimulated sequentially based on the motion phases.
- **3. Active training**: the robot is controlled to follow the patient's intention estimated by sEMG signals.



Training Mode

——FES-Assisted Training

Video – FES-assisted cycling and pedaling movement



Lower-limb Rehabilitation Robot

—Rehabilitation Training Mode

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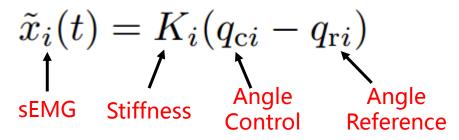


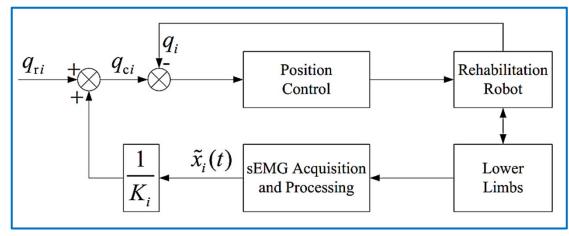
sEMG Sensor

sEMG-Based Active Training Mode

—Spring-type Impedance Interaction Control

- The robot's knee joint angle trajectory is adjusted according to the voluntary contraction of the patient's two thigh muscles.
- Spring-type training controller:

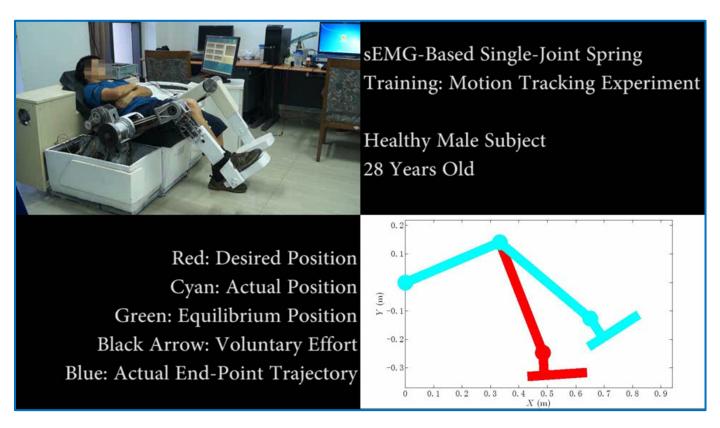




Spring-type Active Training Control Loop

Experiment

—Spring-type Active Training Mode

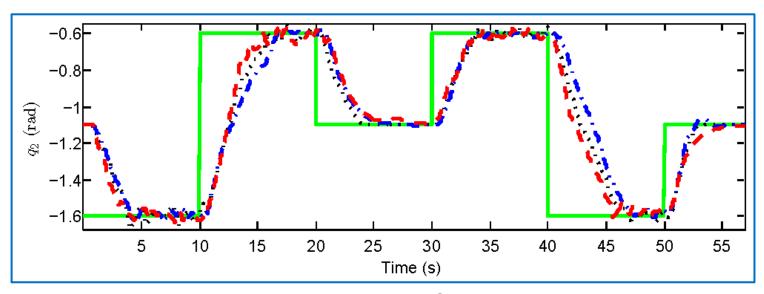


Spring-type Active Training Experiment

Experiment Result

—Spring-type Active Training Mode

Motion Tracking



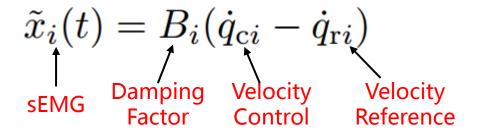
RMS Error (Last 3 Seconds of Each Step)

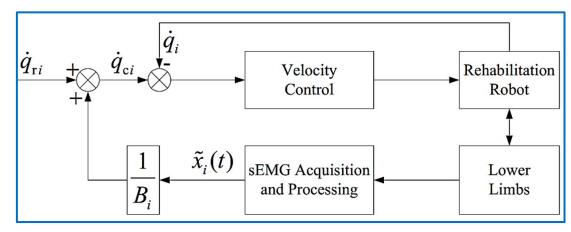
Subject 1	Subject 2	Subject 3
$0.0200 \mathrm{rad}$	$0.0155 \mathrm{rad}$	$0.0237 \mathrm{rad}$

sEMG-Based Active Training Mode

—Damping-type Impedance Interaction Control

- The robot's knee joint velocity trajectory is adjusted according to the voluntary contraction of the patient's two thigh muscles.
- Damping-type training controller:

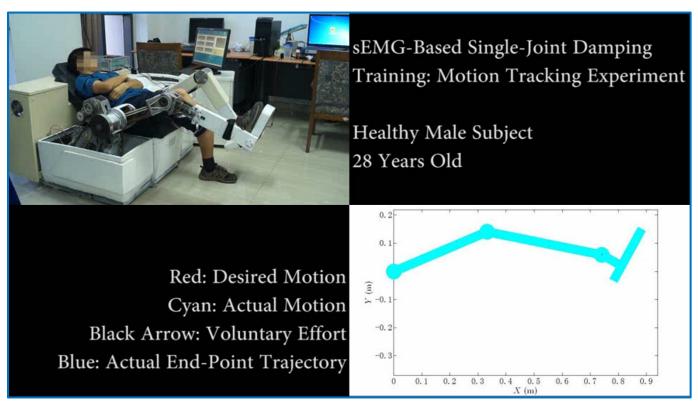




Damping-type Active Training Control Loop

Experiment

—Damping-type Active Training Mode

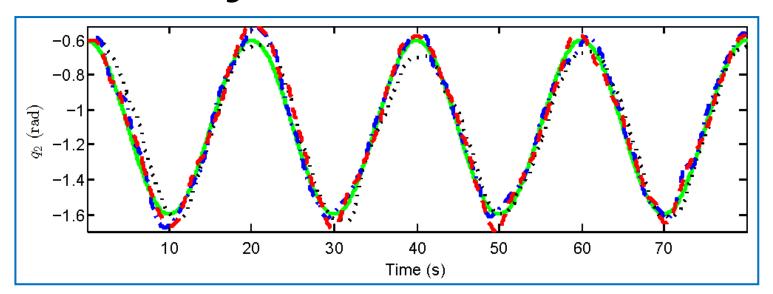


Damping-type Active Training Experiment

Experiment Result

— Damping-type Active Training Mode

Motion Tracking



RMS Tracking Error

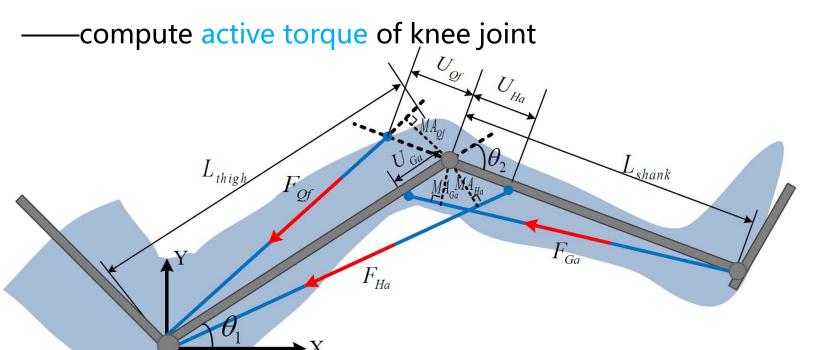
Subject 1	Subject 2	Subject 3
$0.0794 \mathrm{rad}$	$0.0430 \mathrm{rad}$	$0.0483 \mathrm{rad}$

sEMG-Based Active Training Mode

-sEMG-Driven Musculoskeletal Model

2) musculoskeletal model

hip

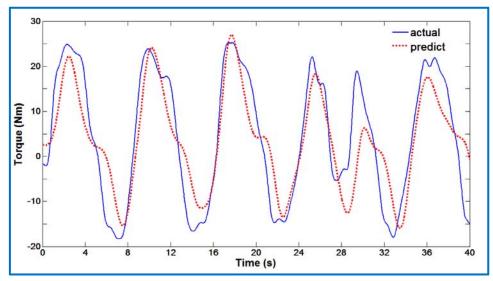


$$T_{act} = \underbrace{K_{Ha} \cdot F_{Ha} \cdot MA_{Ha}}_{Ham strings} + \underbrace{K_{Ga} \cdot F_{Ga} \cdot MA_{Ga}}_{Ga strocnemius} - \underbrace{K_{Qf} \cdot F_{Qf} \cdot MA_{Qf}}_{Quadriceps} + \emptyset(\theta_1, \theta_2)$$

—sEMG-Driven Musculoskeletal Model

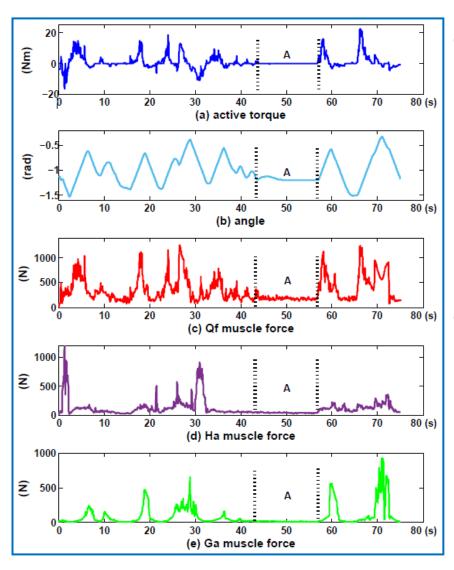
Results of Parameter Identification and Torque Prediction

Subject	U_{Qf} (m)	U_{Ha} (m)	U_{Ga} (m)	K_{Qf}	K_{Ha}	K_{Ga}	<i>RMSE</i> (Nm, Mean ± SD)
1	0.026	0.024	0.006	0.896	0.737	0.909	5.18 ± 1.54
2	0.028	0.018	0.007	0.830	0.983	0.969	5.64 ± 0.83
3	0.021	0.019	0.008	1.194	0.808	0.956	6.05 ± 0.99



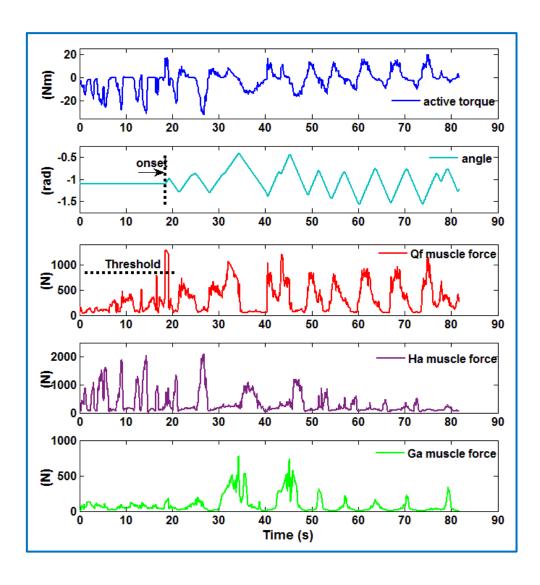
Torque prediction result using the tuned EMG-driven model (RMSE = 7.31 Nm).

Experiment I: Patient-Driven Voluntary Exercise Training



- The knee joint performs an extension movement when the value of active torque is greater than 0, otherwise performs a flexion movement. When the subject relaxes during the time period labelled as "A", the knee joint remains still.
 - Advantages: movement is driven by voluntary effort of the patient's paralyzed limb. Thus, patient's ability could be beneficially exploited to allow a patient to actuate the human-machine system.

Experiment II: Muscle-Triggered Voluntary Exercise Training



- The subject's intention to trigger the onset of the robot action is detected by monitoring muscle force.
 The threshold could be adjusted according to the patient's training needs.
- Advantages: to stimulate specific muscle based on customized requirements. The information, such as trigger time and changes of the selected muscle force, could be useful to make quantitative evaluations.



Training Scenario



Demonstration of Robot Motion Control by sEMG



Thank you for your attention!

Since this is an on-going research, more discussions available on site at the tutorial.